

Water Cycle and Land-Atmosphere Coupling in CFSv2

Paul Dirmeyer

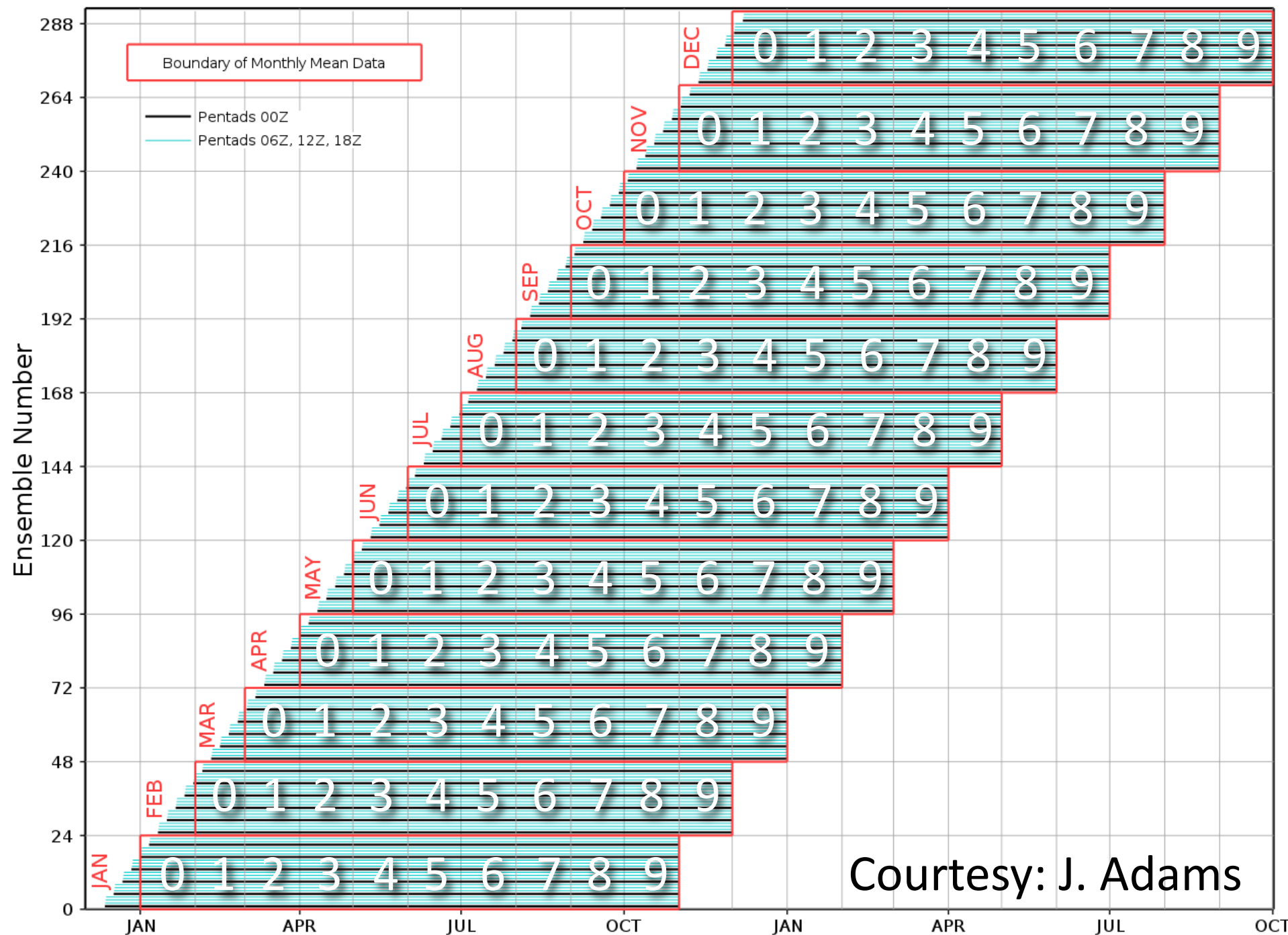
George Mason University /
Center for Ocean-Land-Atmosphere Studies

- Water Cycle Drift in Reforecasts
- Land-Atmosphere Coupling
- Land Initial Conditions and Skill



Another View of the CFSR Suite

Illustration of CFSv2 Reforecast Ensembles



- “0-Lead” starts anywhere from 30 days prior to 7 days into forecast period.
- Four 9-month forecasts started every 5 days.
- 24-28 forecasts per month.
- Creates analysis challenges and opportunities.

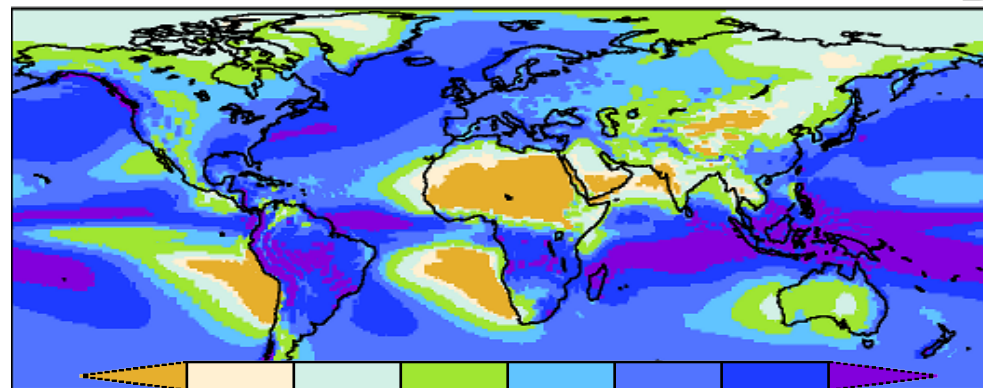


Drift in Means, Variances

- Next slides show seasonal means of:
 - 0-Lead:
 - Mean global precipitation
 - Intra-ensemble standard deviation (24-28 members for each month)
 - Interannual standard deviation (of ensemble means – the “climate signal”)
 - Differences from 0-Lead at leads of 1, 4 and 8 months.
- Quantities averaged for DJF and JJA seasons.

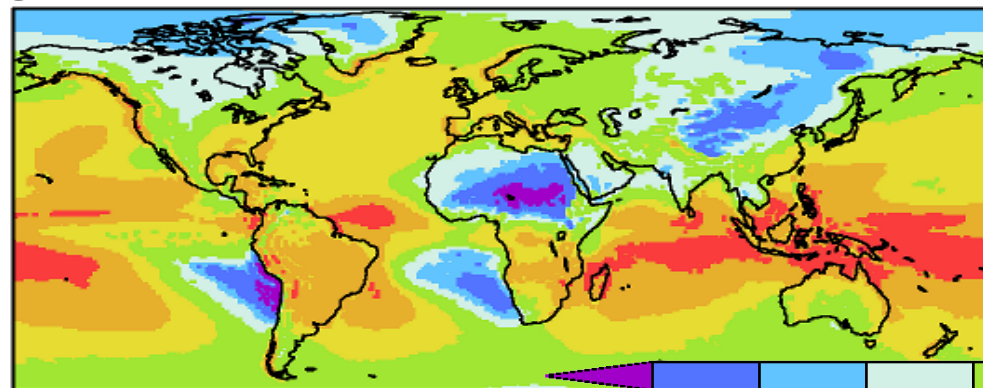


Precip Lead 0

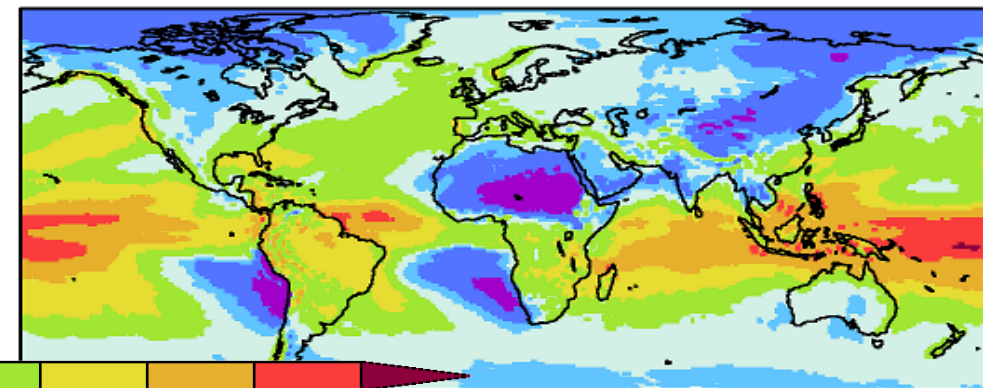


DJF

Intra-Ens_StdDev Lead 0



Interannual_StdDev Lead 0

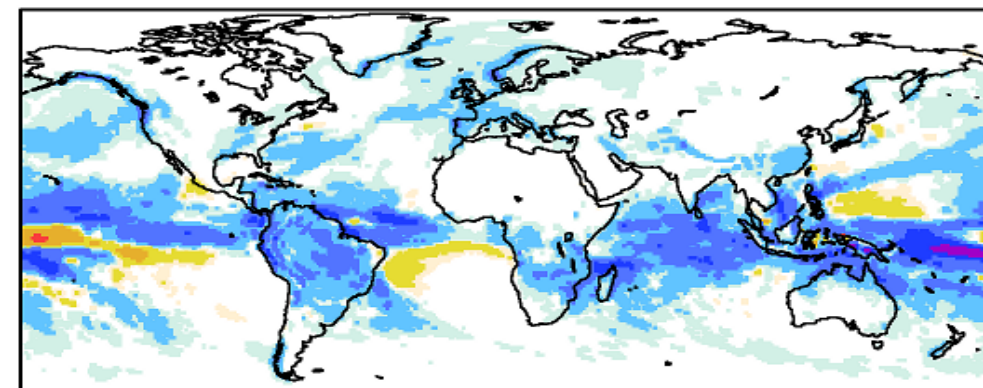
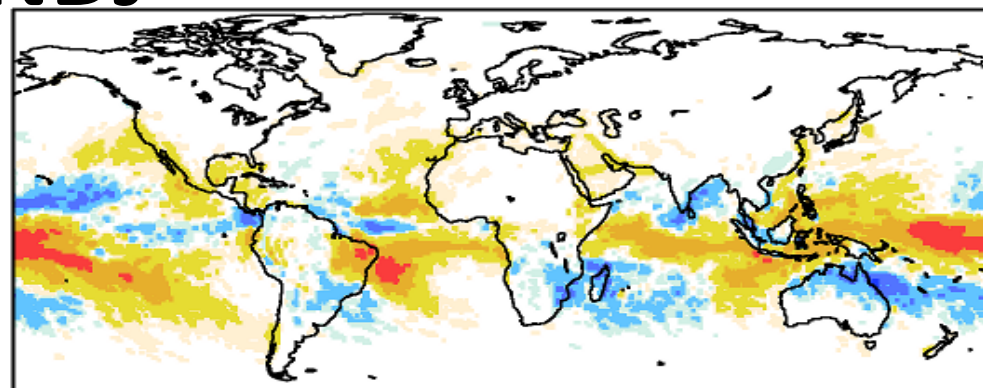
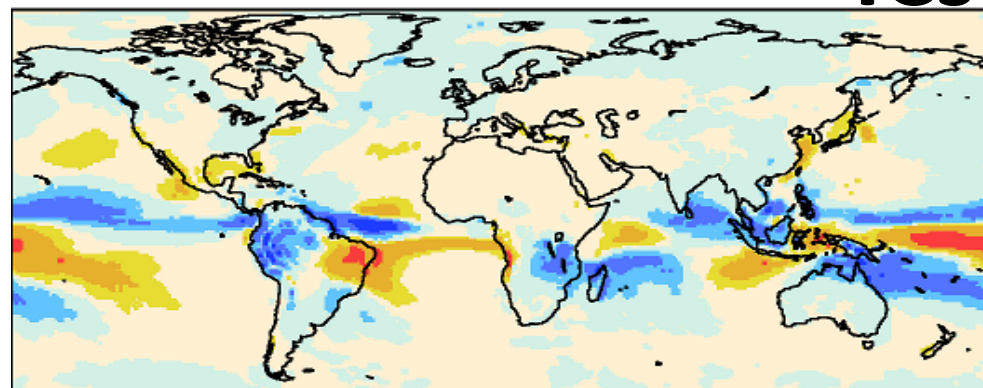


Lead 1

Lead 1

Lead 1

ICs: NDJ

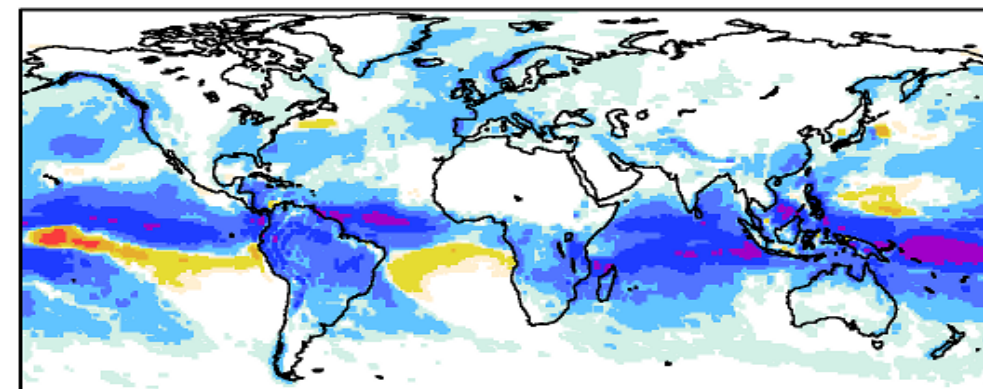
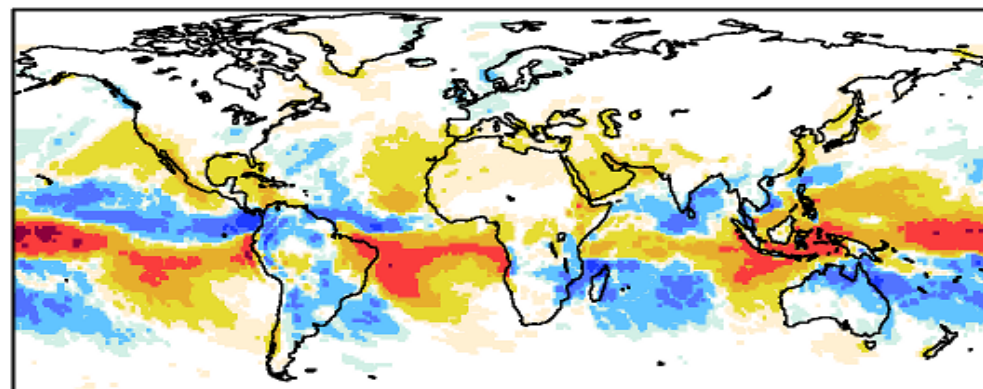
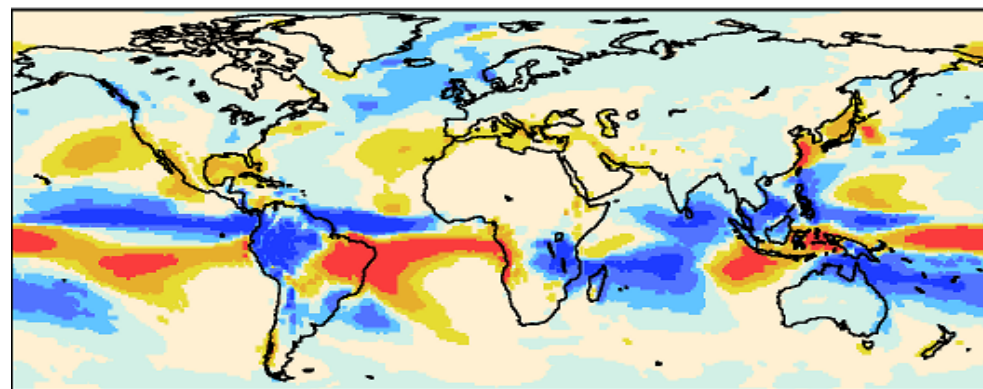


Lead 4

Lead 4

Lead 4

ICs: ASO

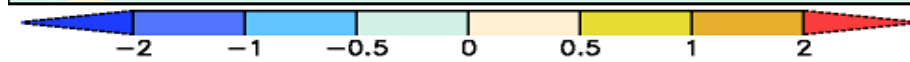
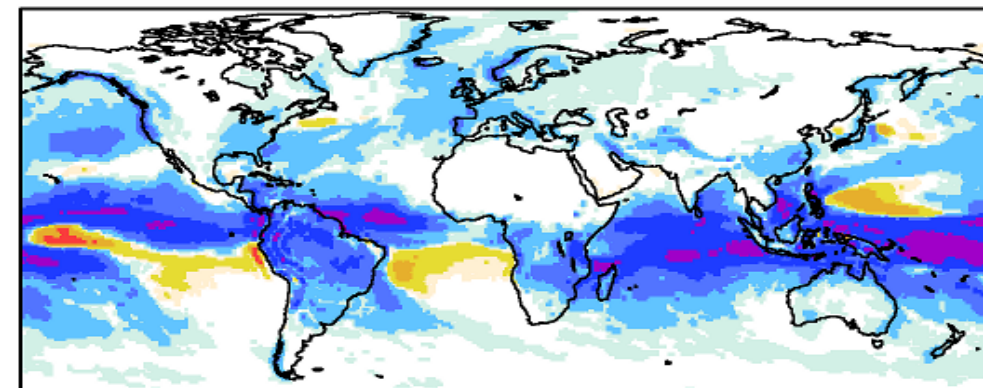
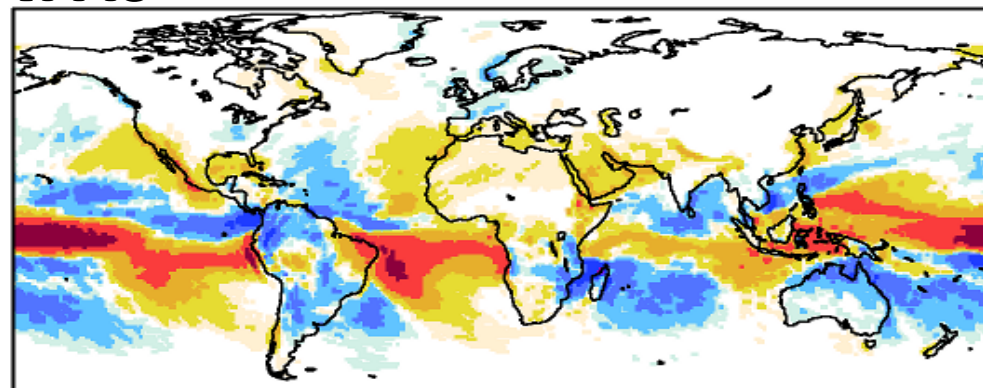
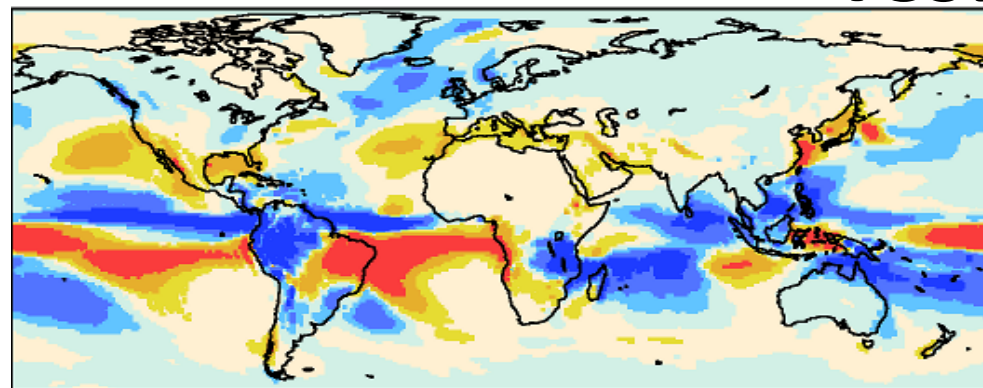


Lead 8

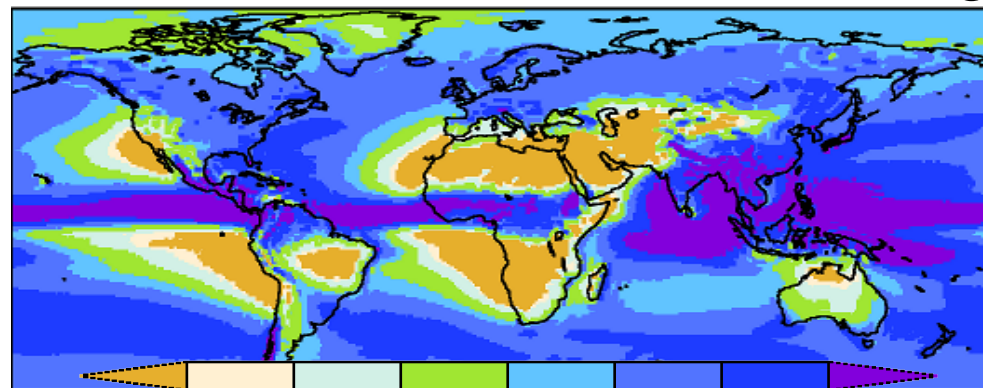
Lead 8

Lead 8

ICs: AMJ

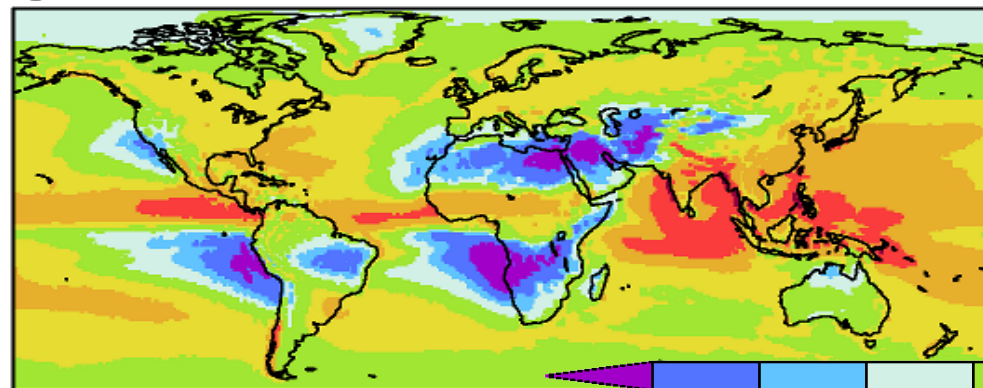


Precip Lead 0

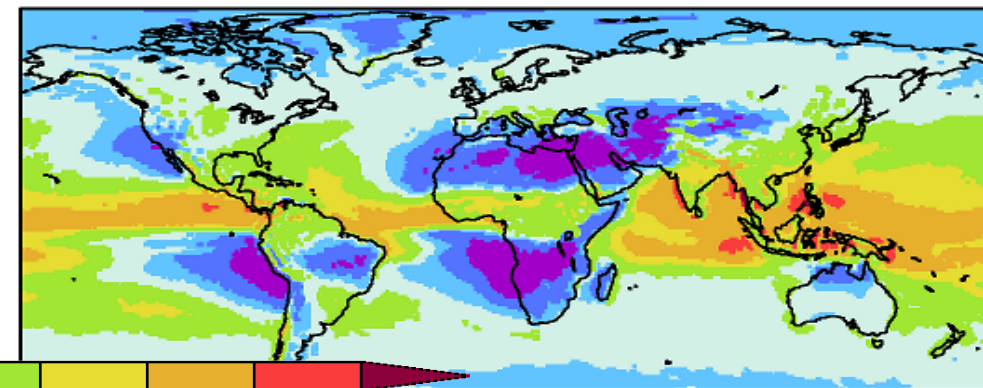


JJA

Intra-Ens_StdDev Lead 0



Interannual_StdDev Lead 0

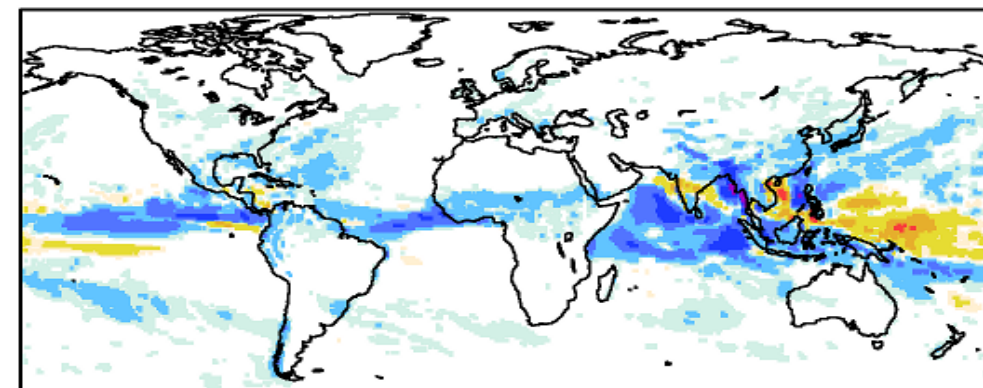
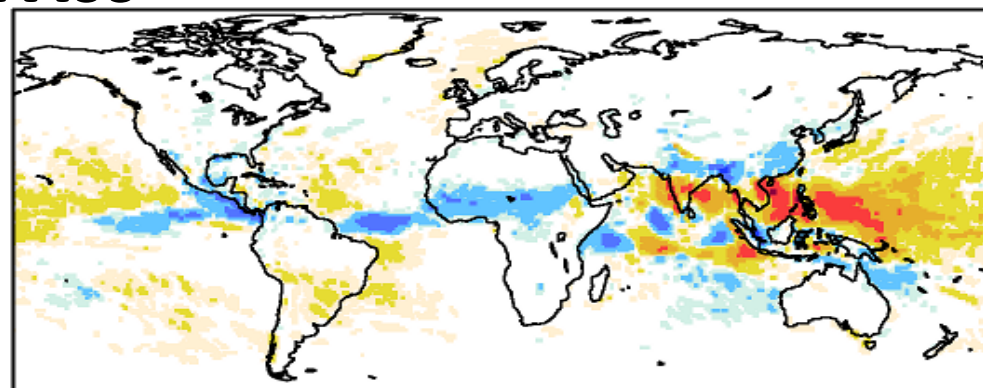
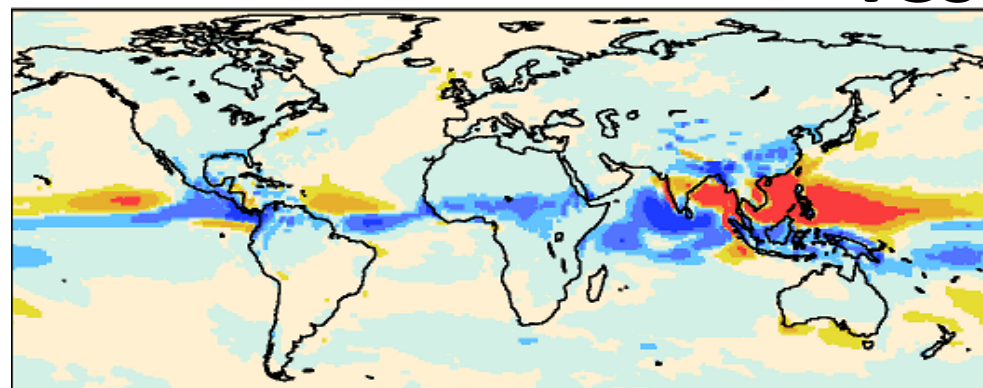


Lead 1

ICs: MJJ

Lead 1

Lead 1

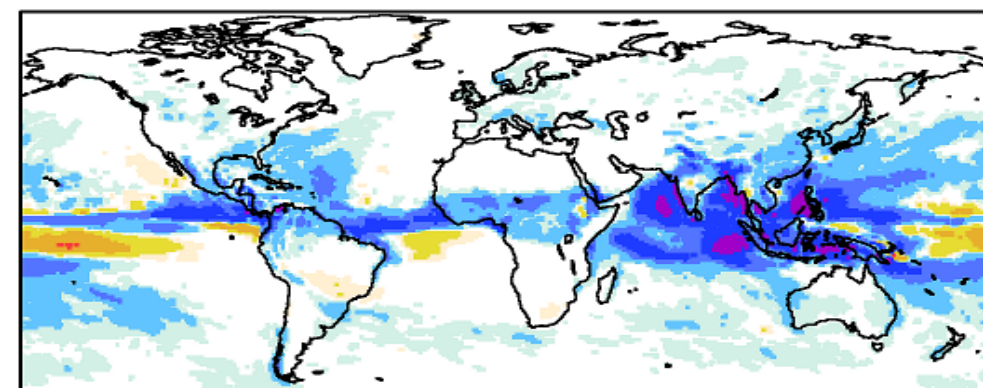
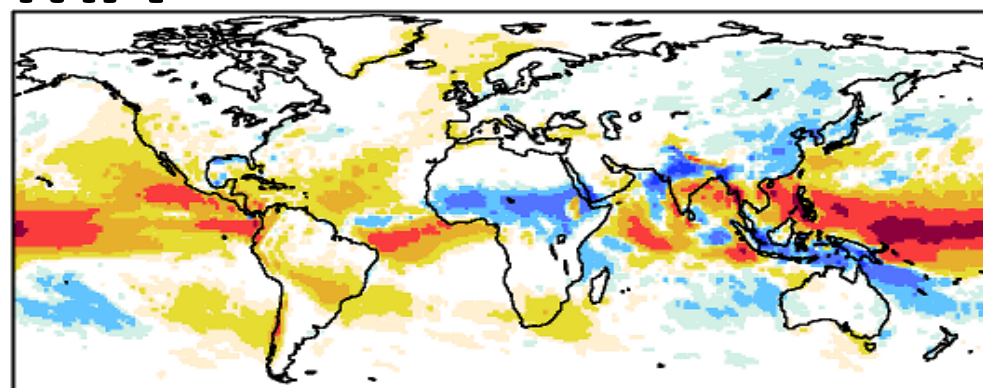
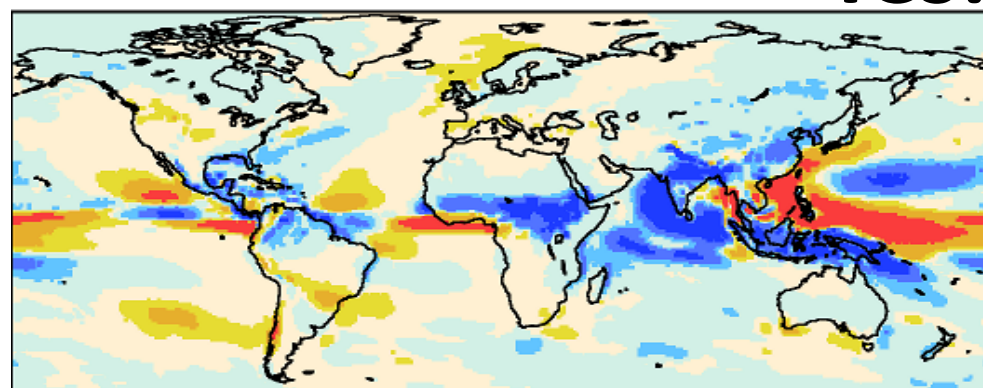


Lead 4

ICs: FMA

Lead 4

Lead 4

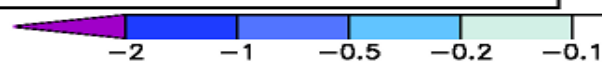
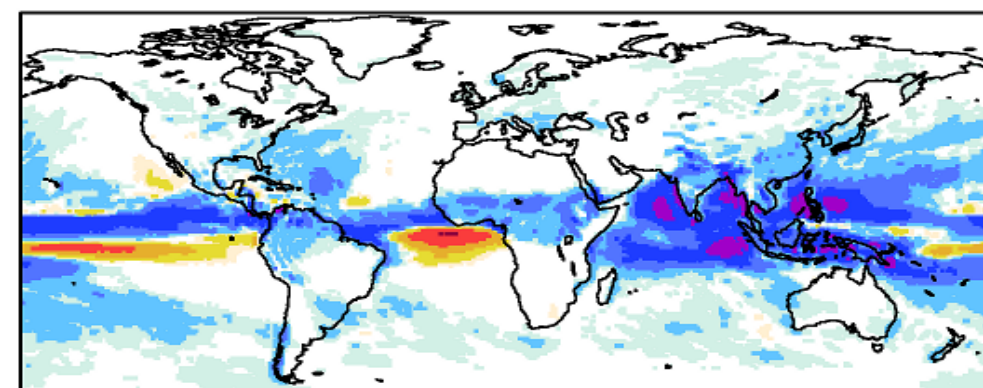
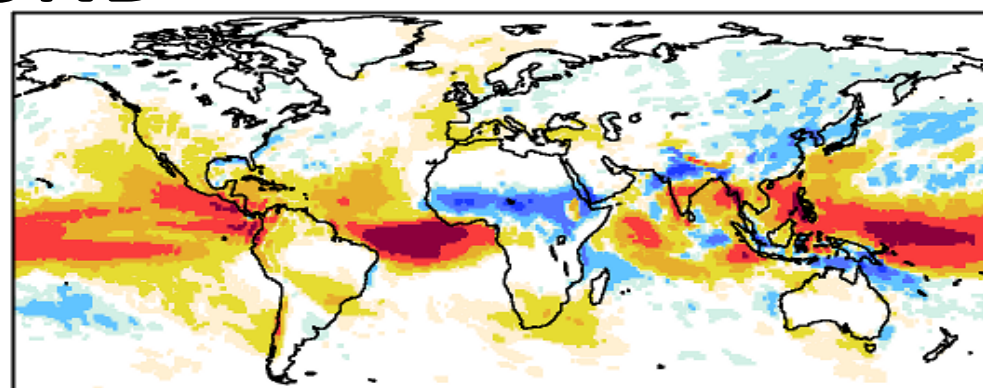
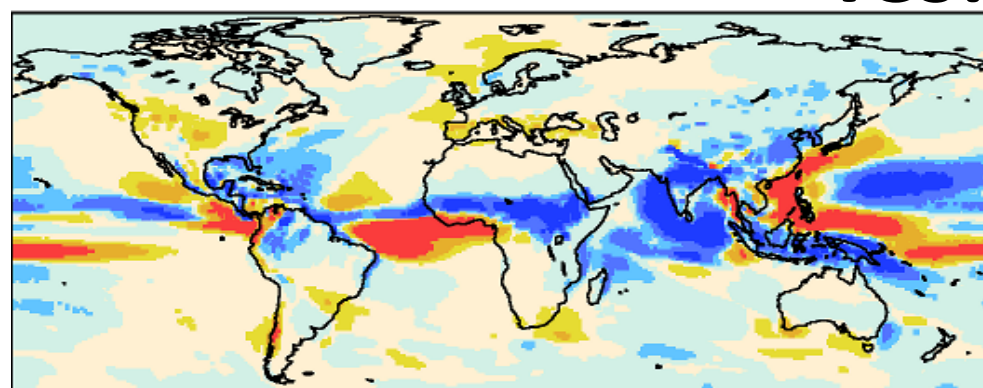


Lead 8

ICs: OND

Lead 8

Lead 8



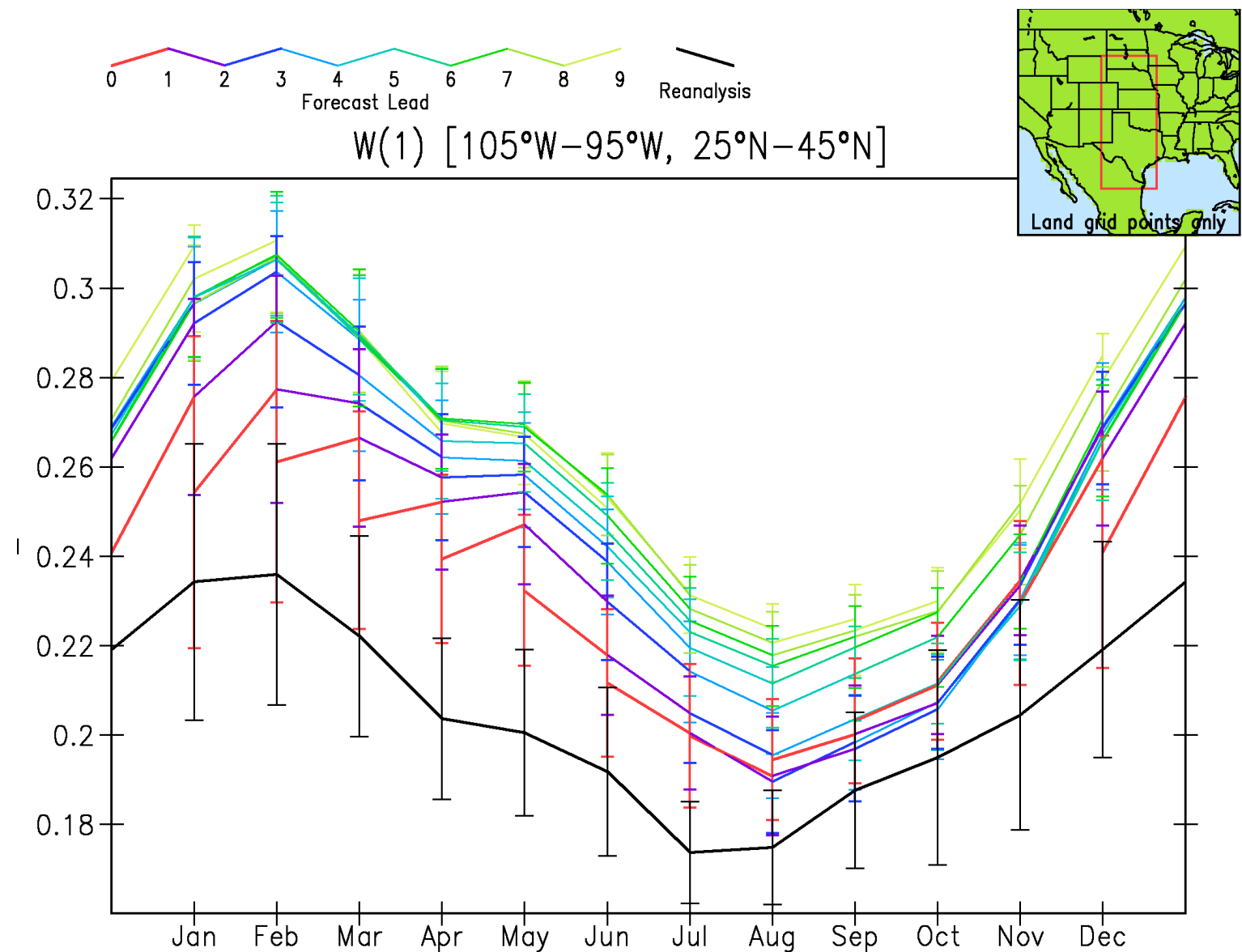
Drift in Means, Variances

- Means show a drift to the CFSv2 model climate
 - Increase in oceanic ITCZ/tropical precipitation
 - Mainly decreases elsewhere (e.g., continents)
- Ensemble spread largely follows mean drift (ocean-driven??).
- Interannual variability decreases markedly with forecast lead.
- Need to consider: **Forecast climatology varies in 2 time dimensions – seasonal cycle and forecast lead**



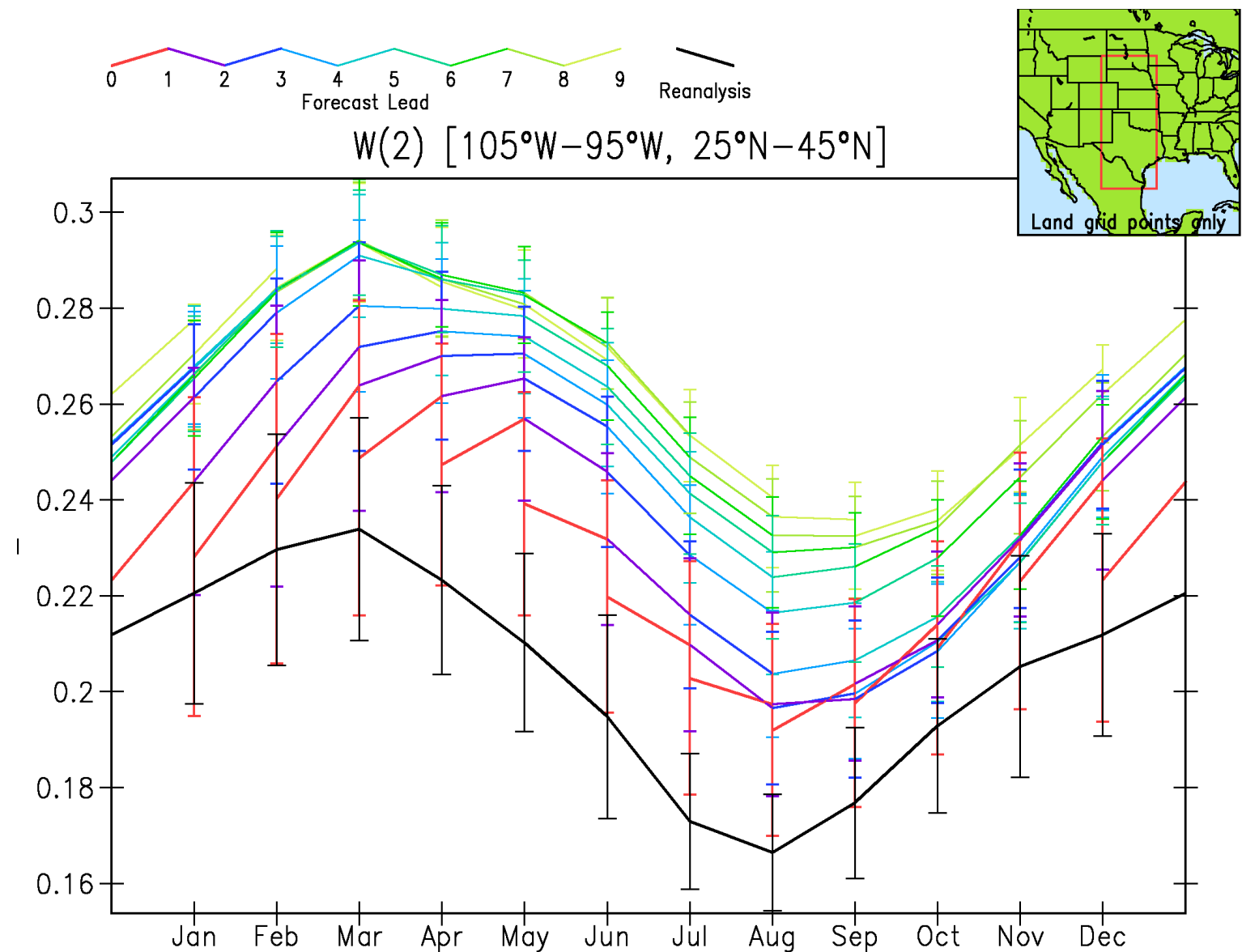
Surface Water Cycle – A Multifarious Tale

- US Great Plains average:
- Surface layer soil moisture quickly drifts towards wet bias
 - Soil moisture initialized from offline Noah run (GLDAS).
 - This is the same stream used to reset CFS reanalysis every 24 hours, so reanalysis (black curve) is constrained by GLDAS.



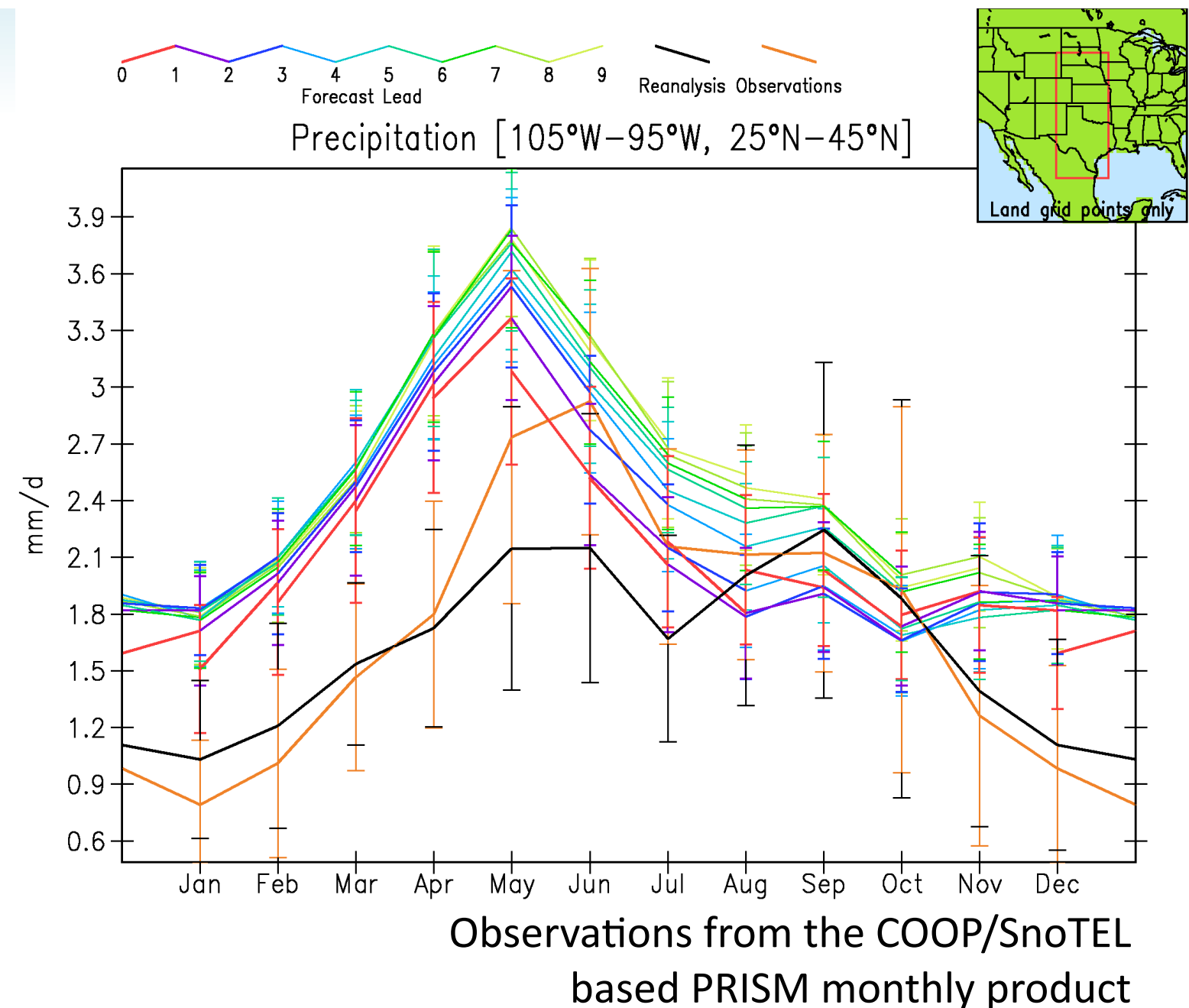
Wet Bias Exists Throughout Upper Soil

- What is the driver of this drift in the CFS reforecasts?
- What are its implications?
- Note – the bars denote the interannual $\pm 1\sigma$ for each forecast at each lead, and for reanalysis.



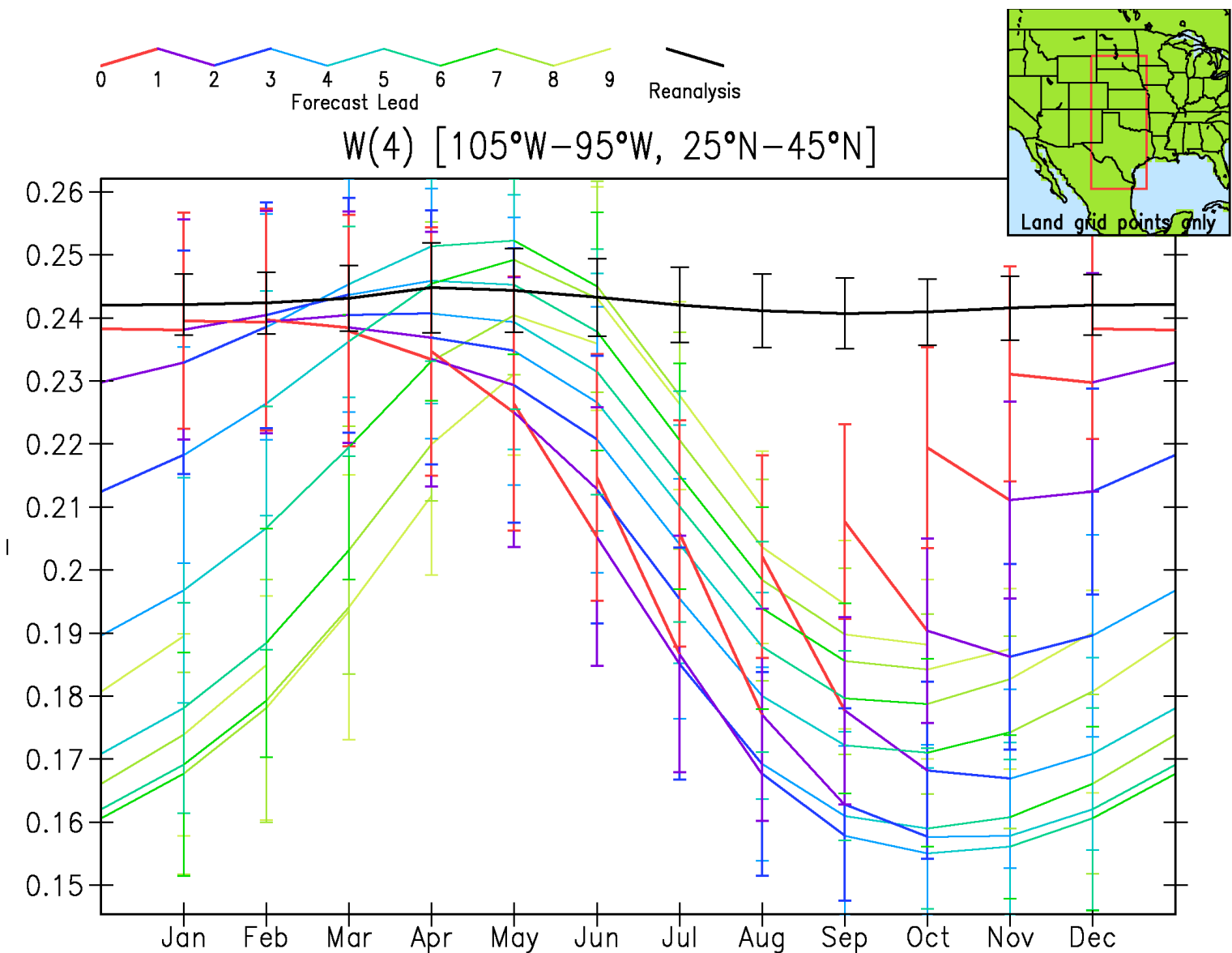
Culprit is Precipitation

- Positive biases in CFSR through much of the year.
 - Reanalysis precipitation actually a bit low in MJJ, yet CFSR is high.
 - Throughout the year, precipitation simulations trend positive with increasing lead time.



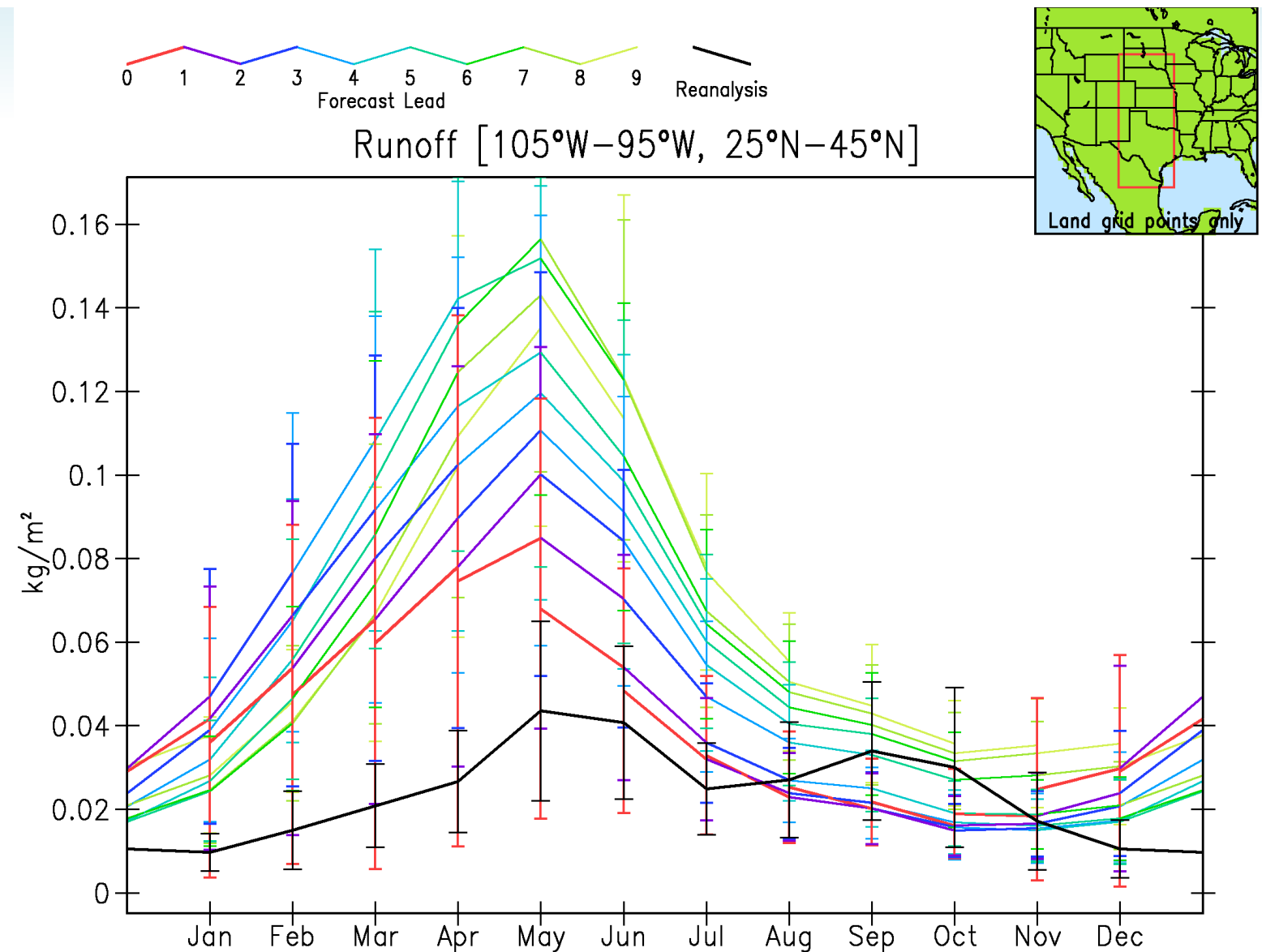
Deep Soil Behaves Quite Differently

- Reanalysis (GLDAS) deep soil moisture is very flat.
- CFSR has a large annual cycle and an oscillation in the evolution of biases.
- Recall reanalysis states are constrained and there is no conservation enforced (characteristic of NWP's DAS).
- The reforecasts are in a model that (largely) closes the water budget.



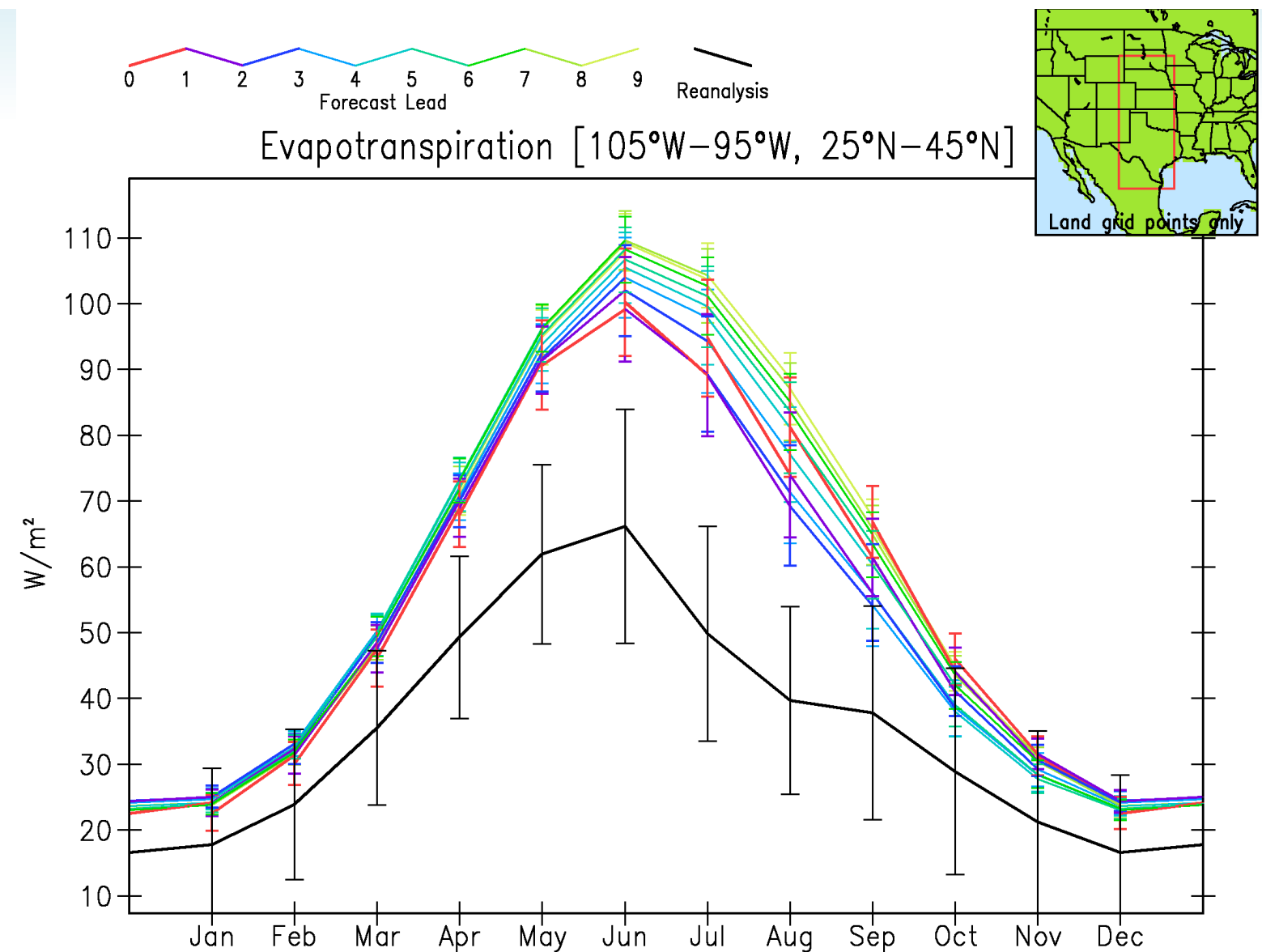
Runoff

- CFSR output does not discern between baseflow and surface runoff, but we see the signature of precipitation biases.
- How can the hydrologic community make use of such data?



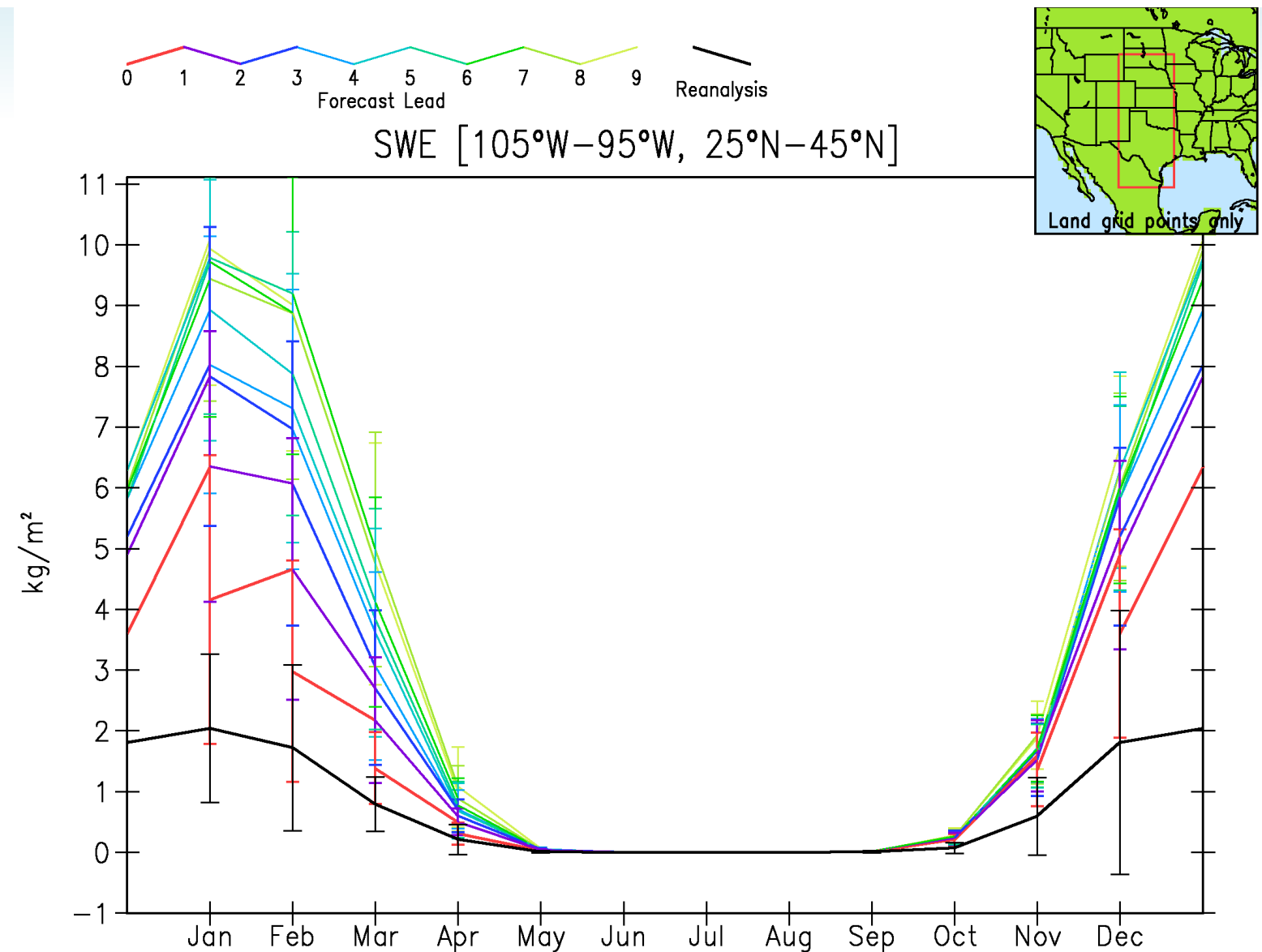
Evapotranspiration

- CFSR ET is very high compared to the reanalysis.
- Recall CFSR SM is higher than in upper soil – source of extra ET.
- **Implication** – negative increments in SM in reanalysis – soil constantly dried, this limits moisture for ET.
- Forecasts free to run up.



Snow – the Final Frontier(?)

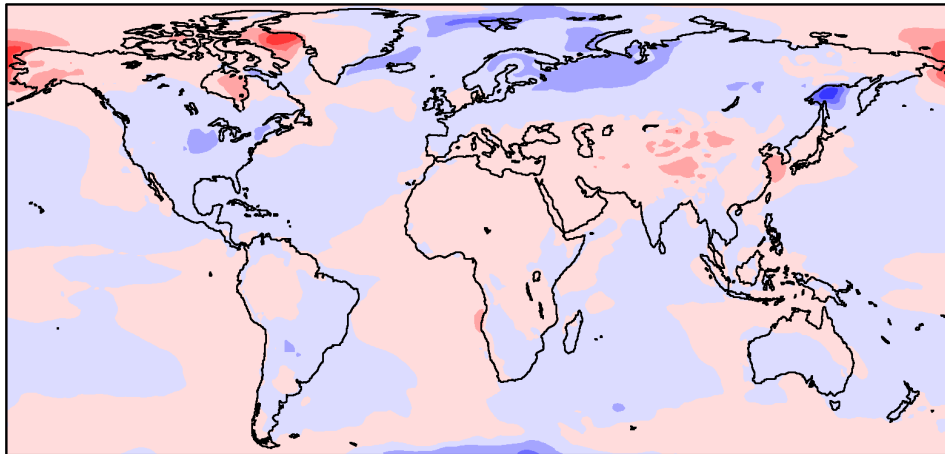
- The largest biases (percentage) appear to be in snowfall / snow cover.
- Biases also across North America and Eurasia.
- All of these evolving biases in water budget terms pose challenges to users in hydrologic, agricultural, and related fields.



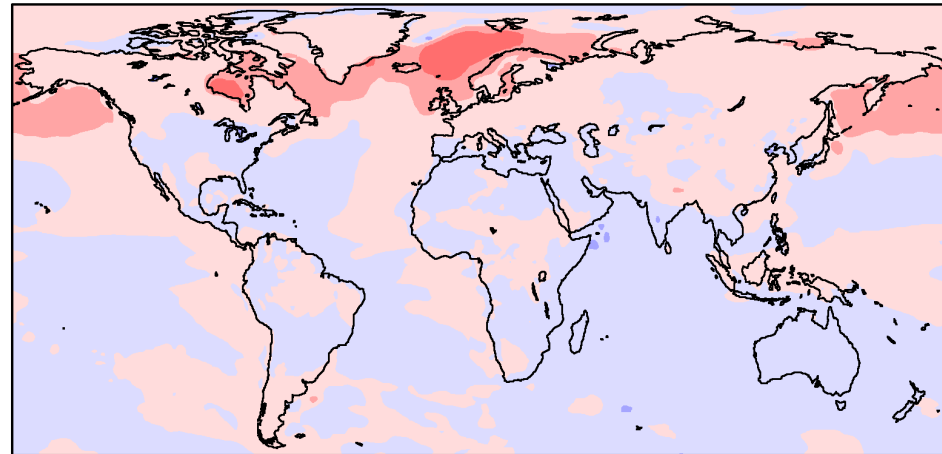
Snow Bias Signal in Temperature

Lead-0 to Lead-1 Drift in Temperature

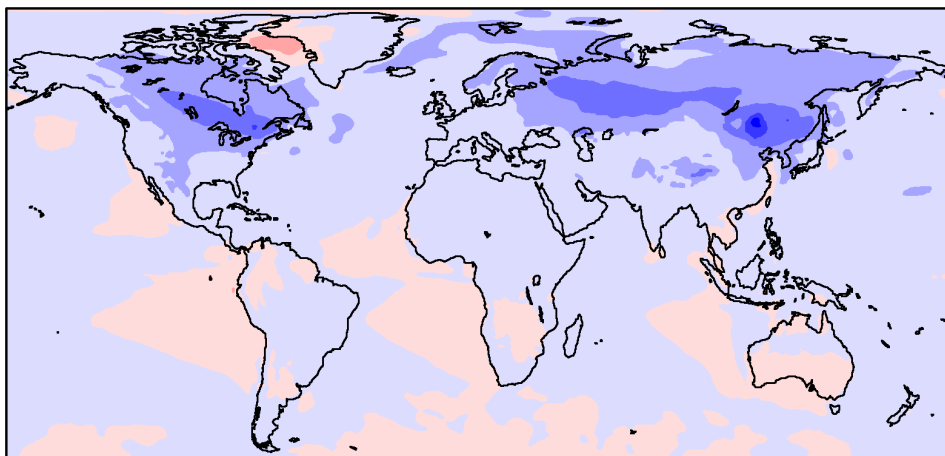
January



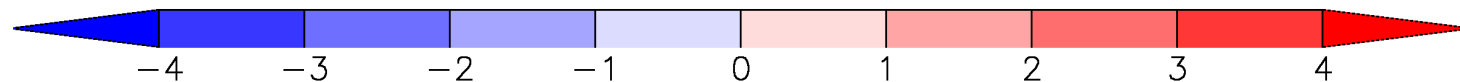
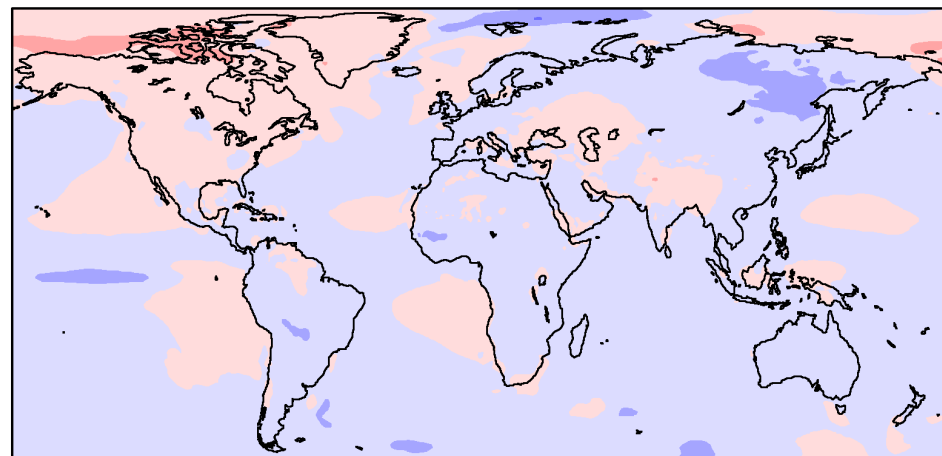
July



April



October

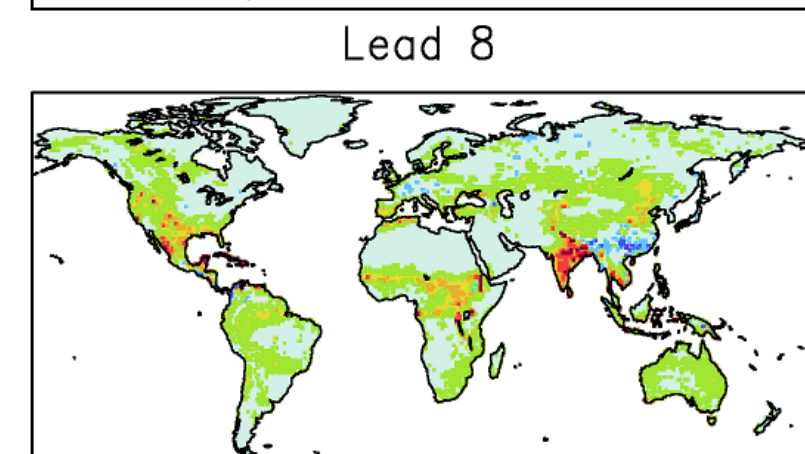
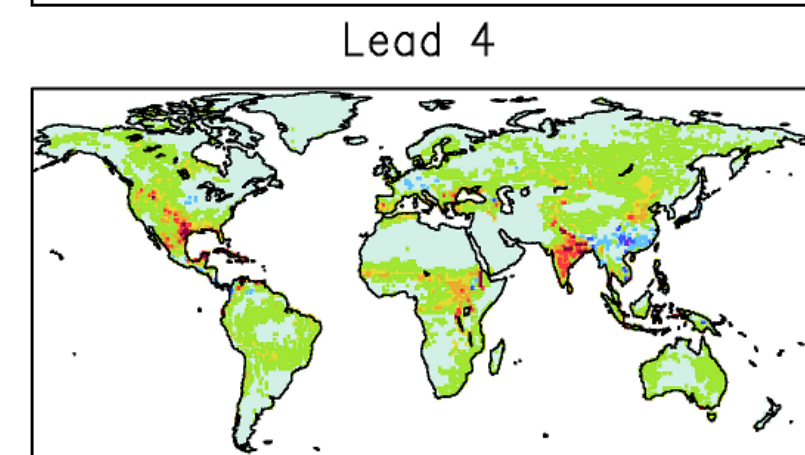
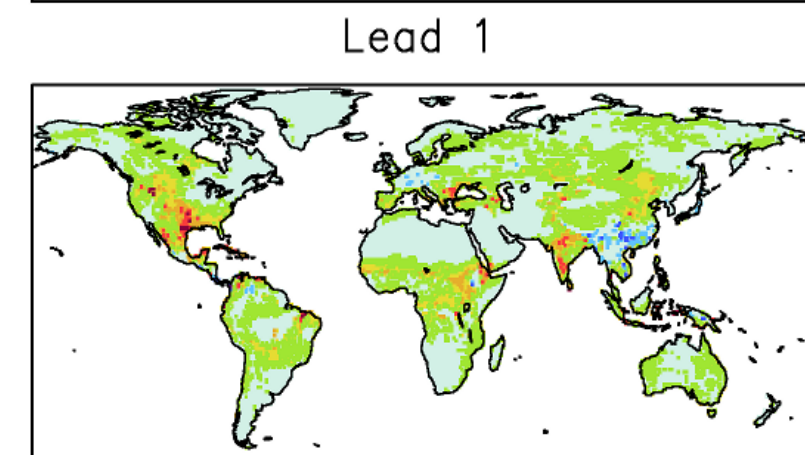
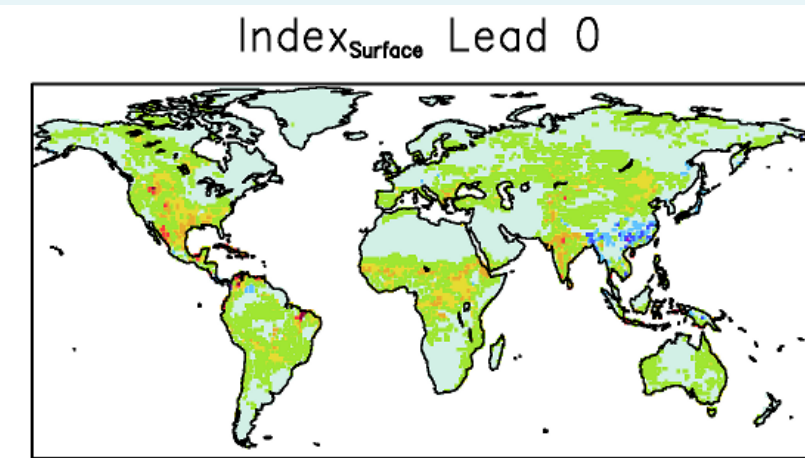
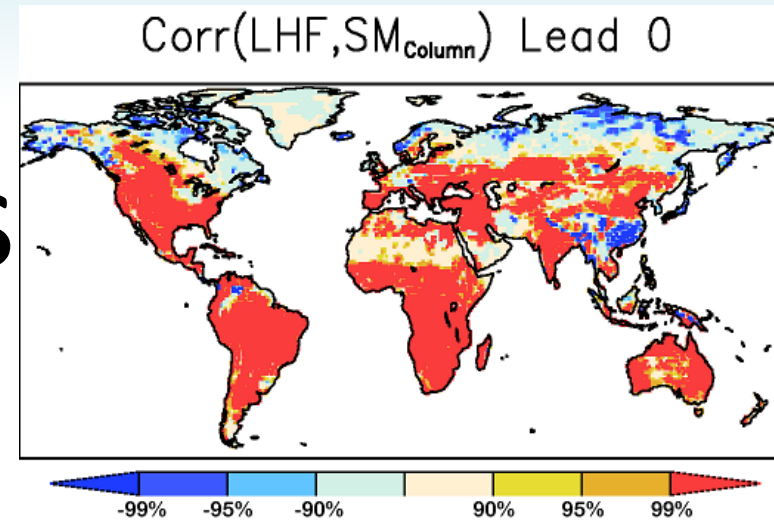


- Heavy snow bias and late snow melt manifests as cold bias in spring.
- Evident in many other states and fluxes as well.



July Coupling Indices

- Positive correlation between evaporation and soil moisture indicates soil moisture is controlling surface fluxes.
 - Necessary condition for feedback
- Index is product of $r_{\text{LHF,SM}}$ and σ_{LHF} .
- Index grows with lead over US (spring ICs) and India (winter-early spring ICs)
 - Indicative of systematic precip errors.



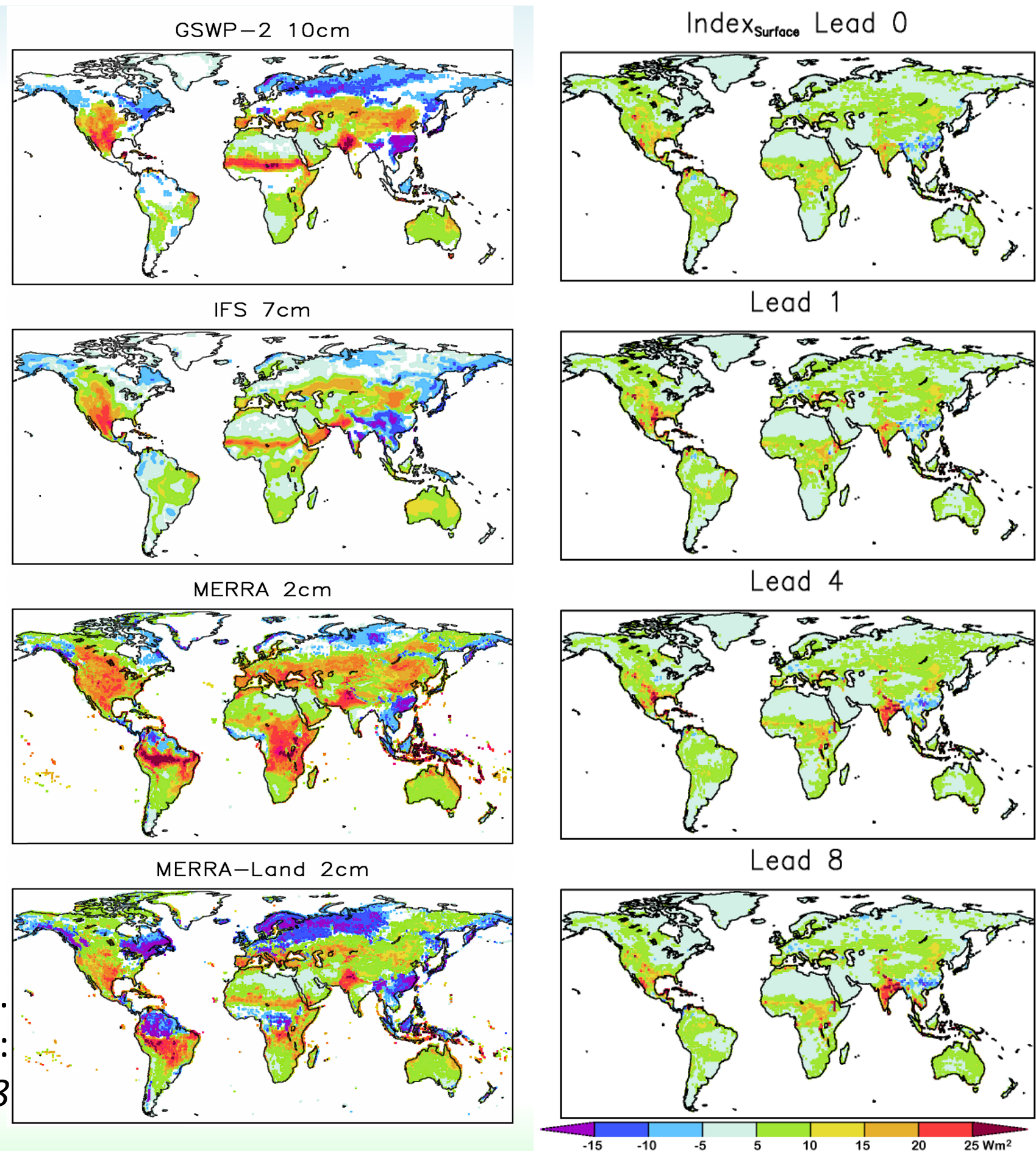
How Does CFSv2 Compare?

- Index for CFSv2 with Noah is considerably weaker (+&-) than:
 - GSWP-2 (Land MME)
 - IFS run in climate mode
 - MERRA reanalysis (both L-A and the land-only “replay”).

Still July....

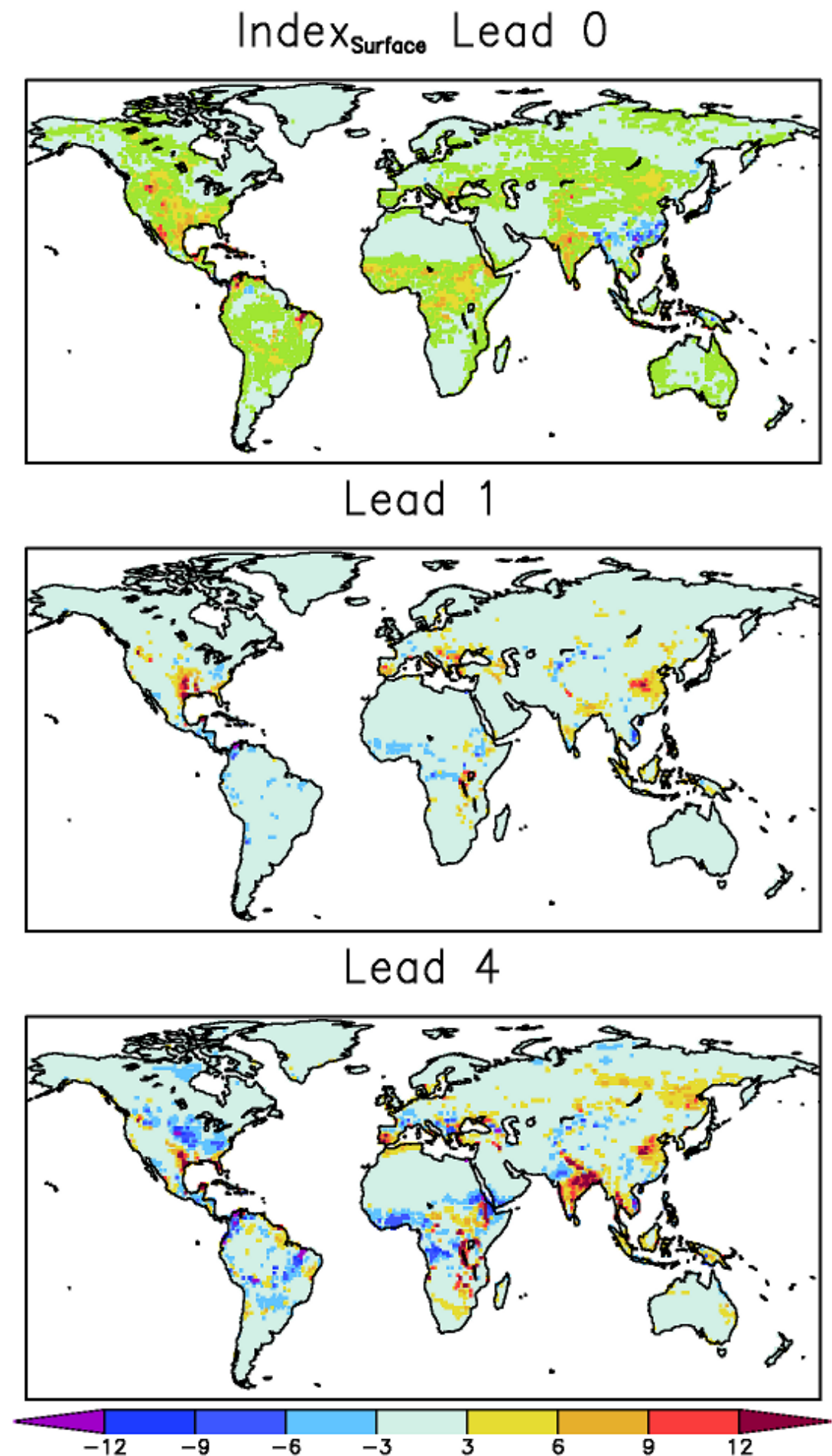
Left panels from Dirmeyer (2011):
GRL doi:10.1029/2011GL048268

1 May 2012

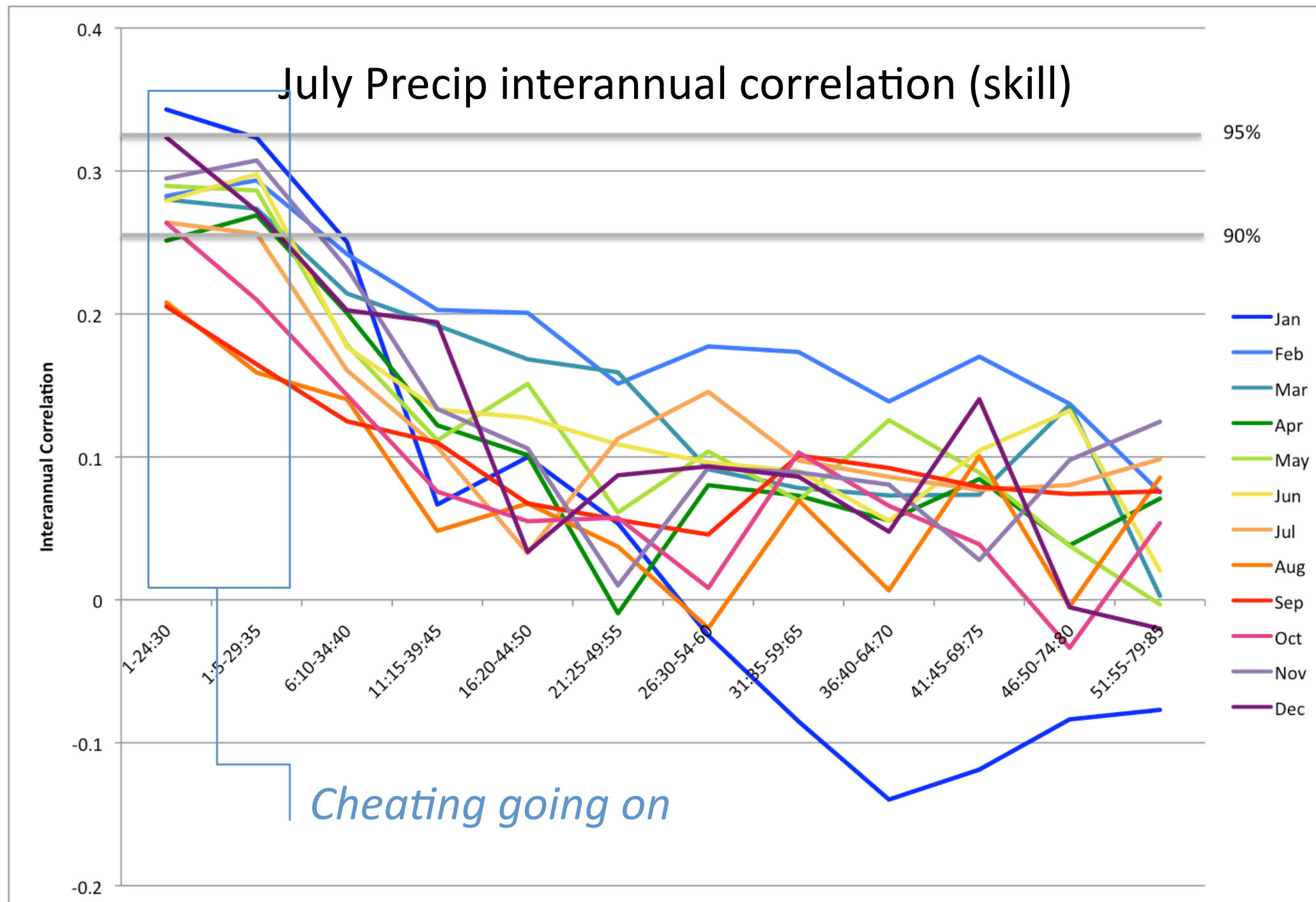


Drift in Coupling

- Changes in coupling index shows the southern Great Plains gets stronger, but much of the rest of North America has weakening coupling.
- These changes come because soil moisture drifts in/out of “sweet spot” for flux sensitivity.
- Could this contribute to reduced skill (cf GLACE-2)?



Precipitation Skill by Lead

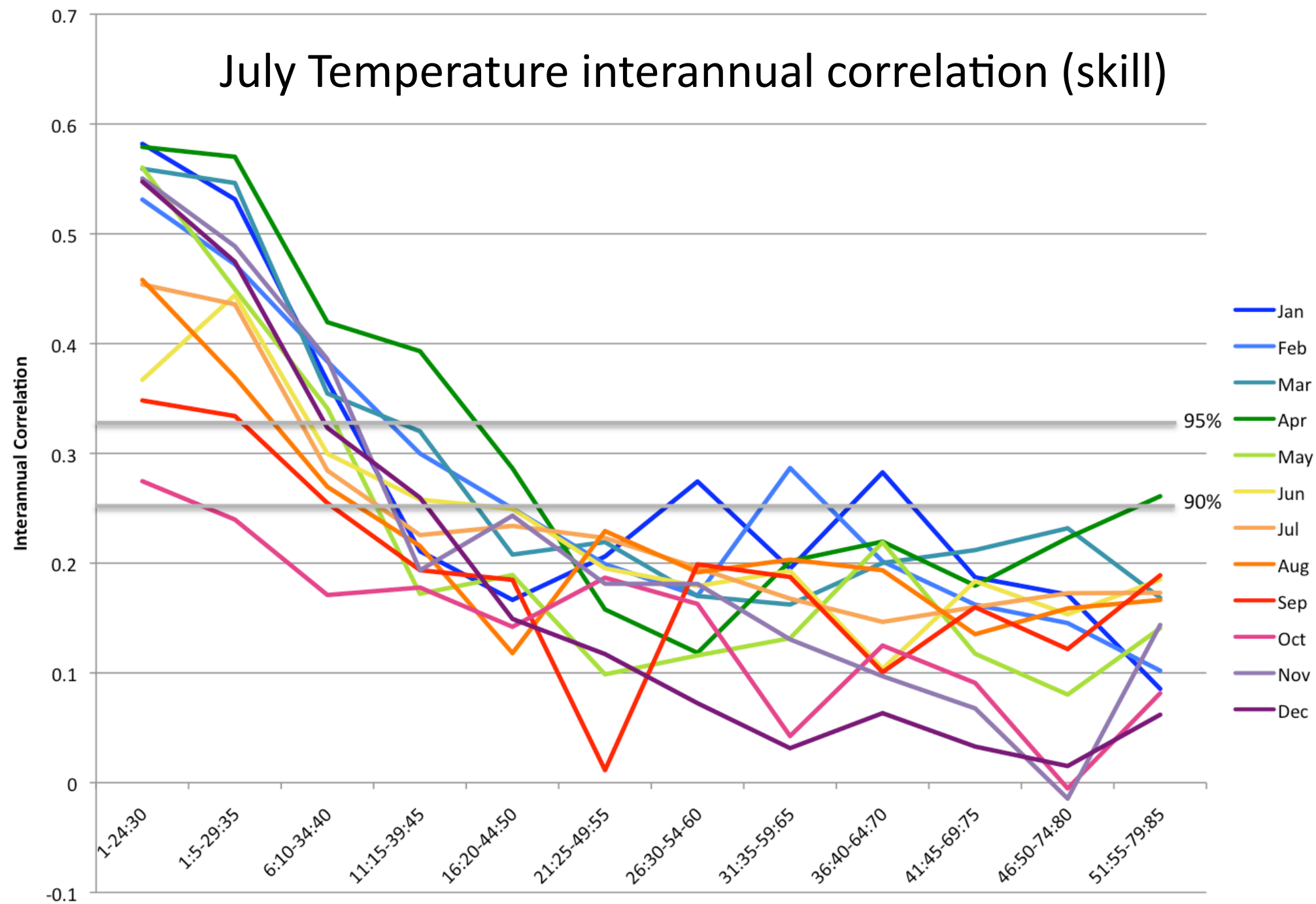


- Staggering of ICs in ensembles allows for a pentad-level assessment of skill.
- Averaged over CONUS, little skill in monthly means.



Temperature is Better

July Temperature interannual correlation (skill)

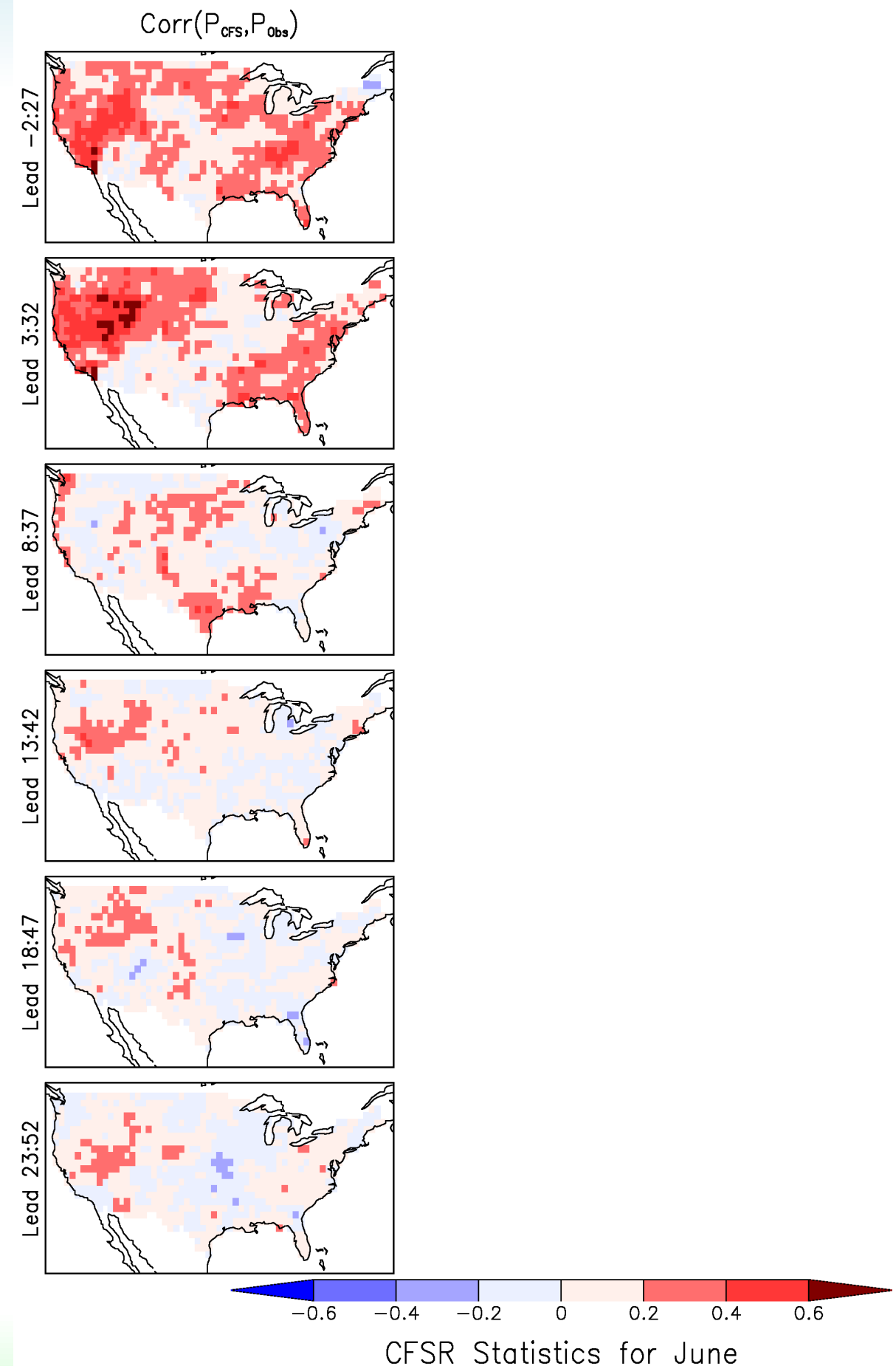


- Have yet to look at seasonal (3 month) skills – likely to be better than single month skills for both precipitation and temperature



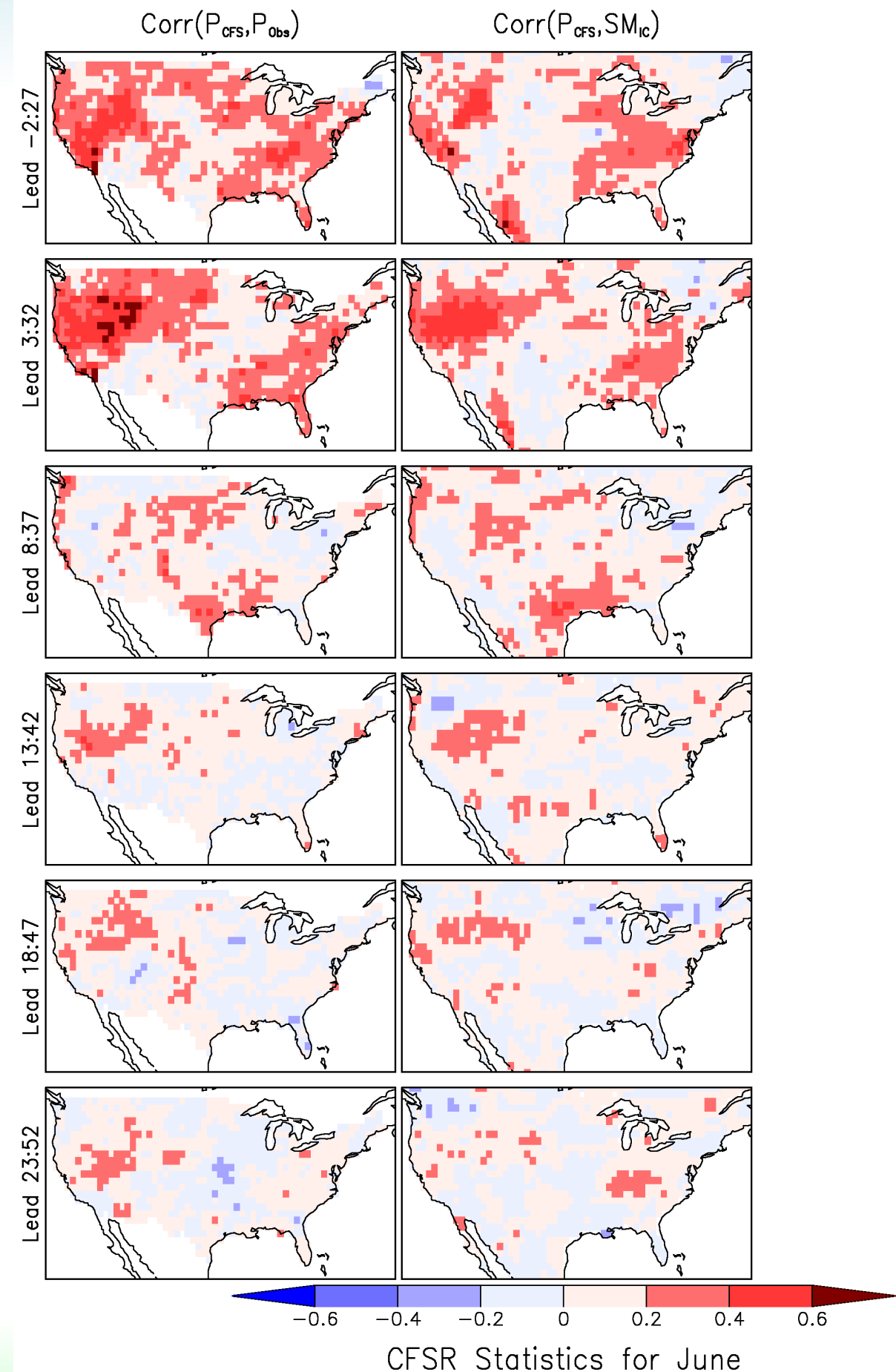
Precipitation Validation and Soil Moisture ICs

- CFSR monthly precipitation rapidly decorrelates from obs.



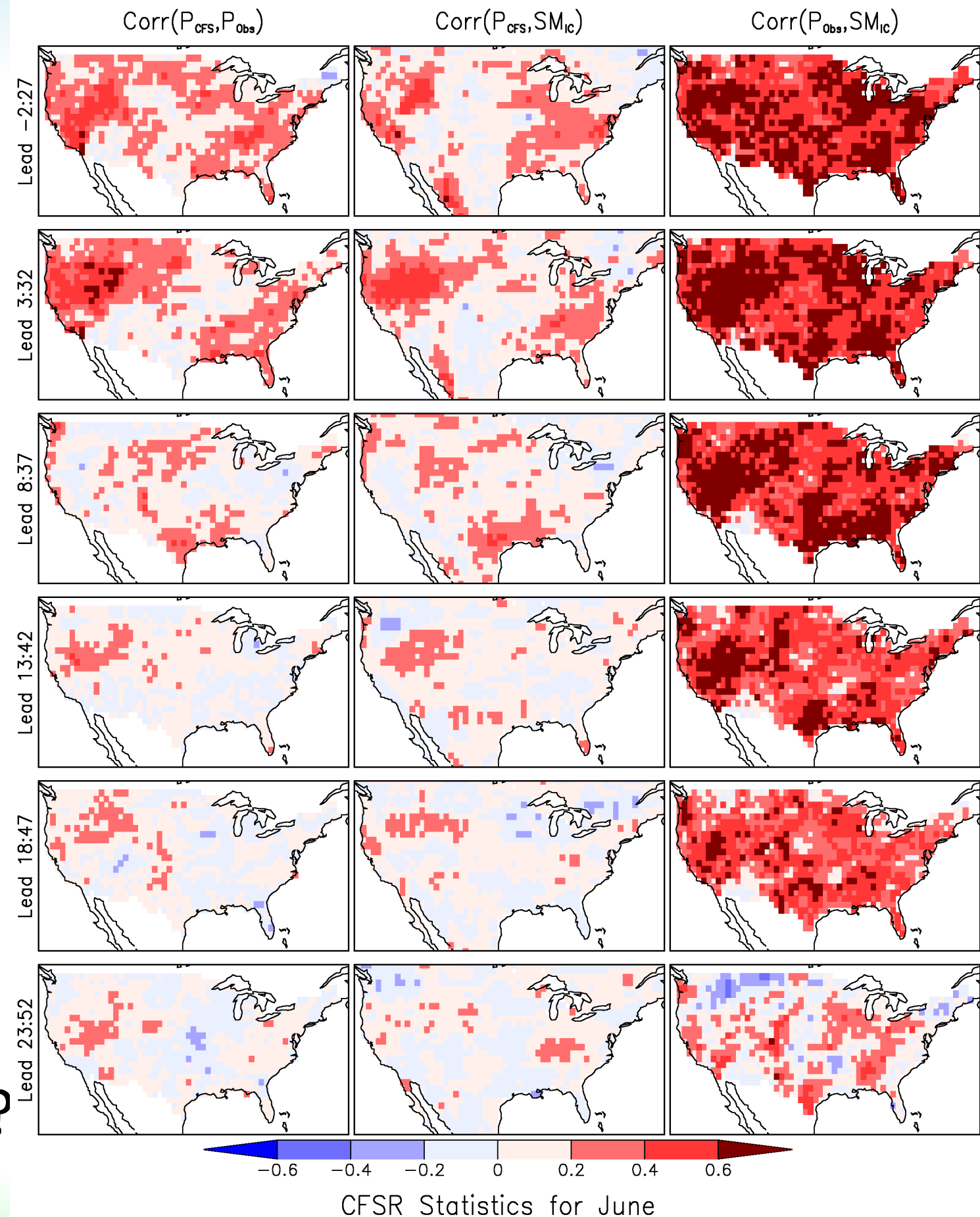
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- CFSR precip similarly loses correlation with initial surface soil moisture anomalies.



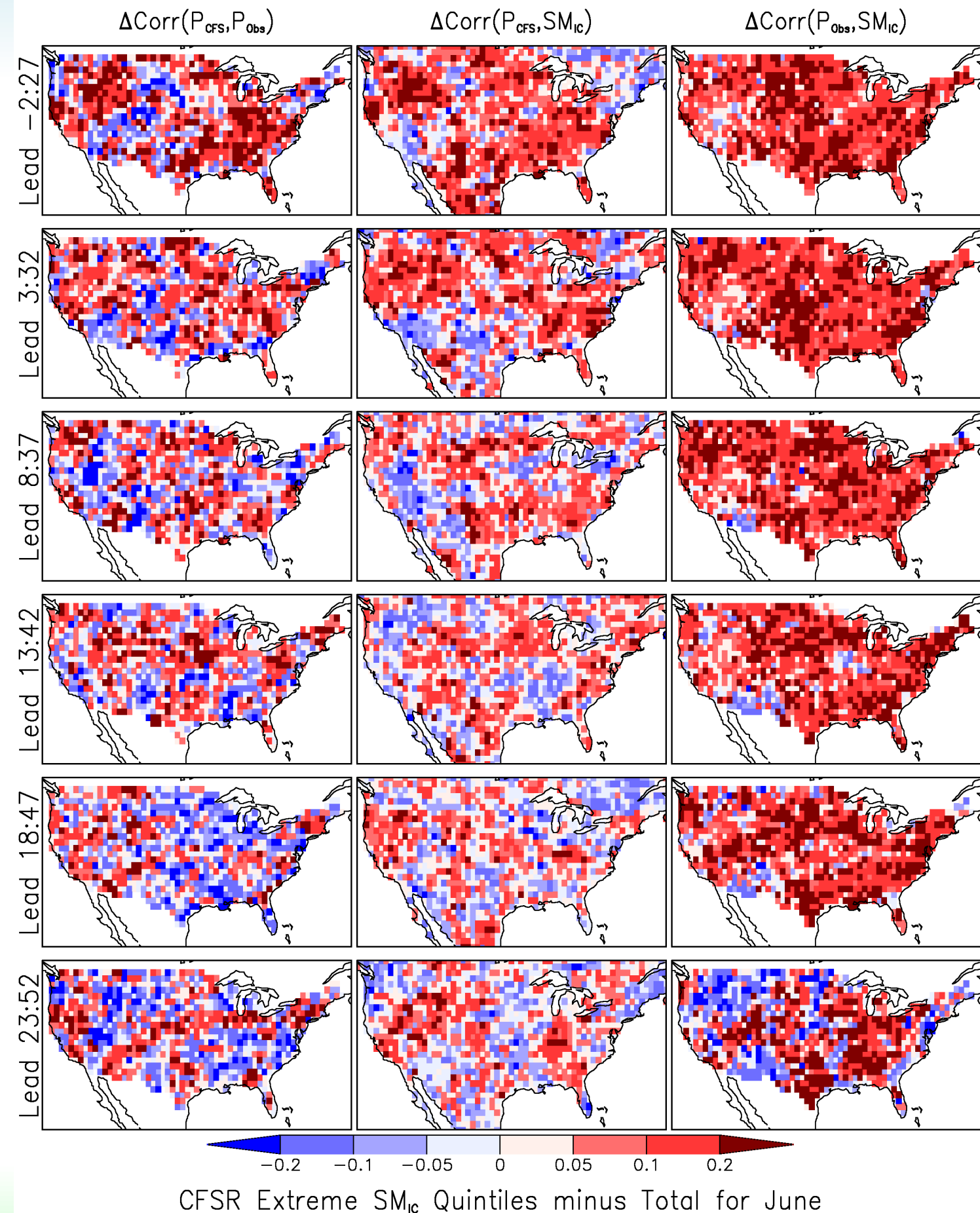
Precipitation Validation and Soil Moisture ICs

- CFSR monthly precipitation rapidly decorrelates from obs.
- CFSR precip similarly loses correlation with initial surface soil moisture anomalies.
- Observed precipitation has **much stronger** correlation with antecedent soil.
- Why? Positive L-A feedback, or persistent weather regimes?



And Extremes?

- Plots show the changes in correlation when only the forecasts with the driest/wettest 20% of soil moisture ICs are used (compared to previous slide).
- More skill and connection of forecasts to SM_{IC} .
- Observations also show even stronger correlations.
- Still an open question: what is the cause?



Summary

- Huge drifts exist – CFSv2 climate is not naturally near the CFS Reanalysis climate. CFSv2 climatology varies in 2 time dimensions.
- Drifts and increments in state variables affect fluxes – this is very evident in local/regional water budgets.
- Land-atmosphere coupling metrics show patterns in good agreement with other global estimates, but generally weaker.
- Subsequent rainfall too weakly correlated with antecedent soil moisture
 - How much is weak coupling and how much is excessive high-frequency variability (or is variance simply following excessive mean rainfall)?

